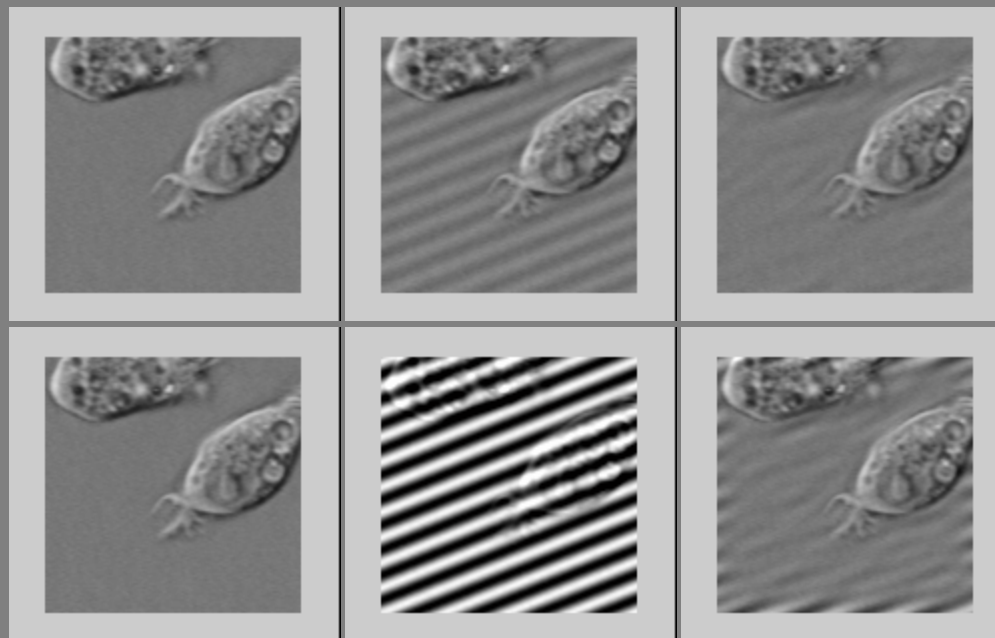


Image Processing and Analysis II



Materials extracted from Gonzalez & Wood
and Castleman

Image Preprocessing – Morphological Operations I

Let A and B be a set of points in the image.

Translation of A by a vector x: $(A)_x = \{c \mid c = a + x, \text{ for } a \in A\}$

Reflection of B: $\hat{B} = \{x \mid x = -b, \text{ for } b \in B\}$

Complement of A: $A^c = \{x \mid x \notin A\}$

Difference of A & B: $A - B = \{x \mid x \in A \text{ and } x \notin B\}$

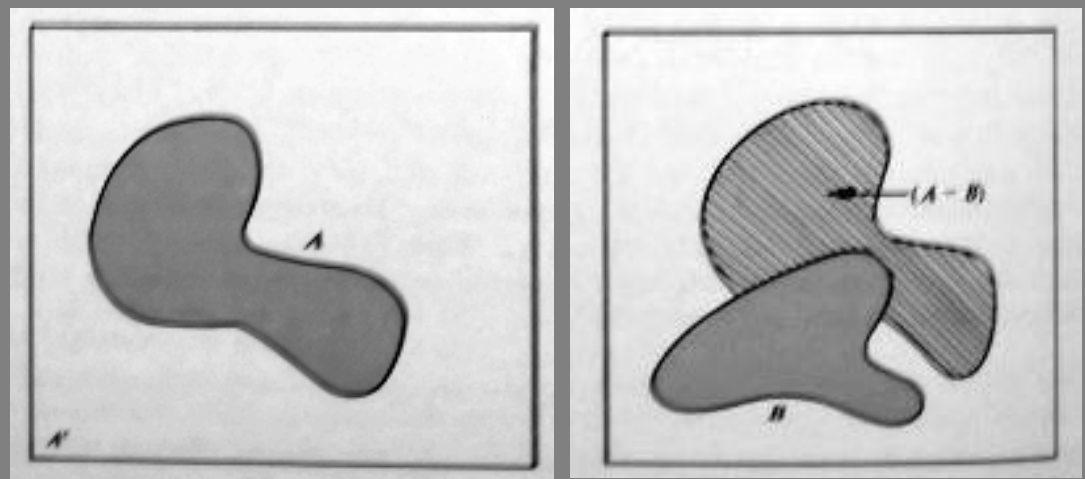
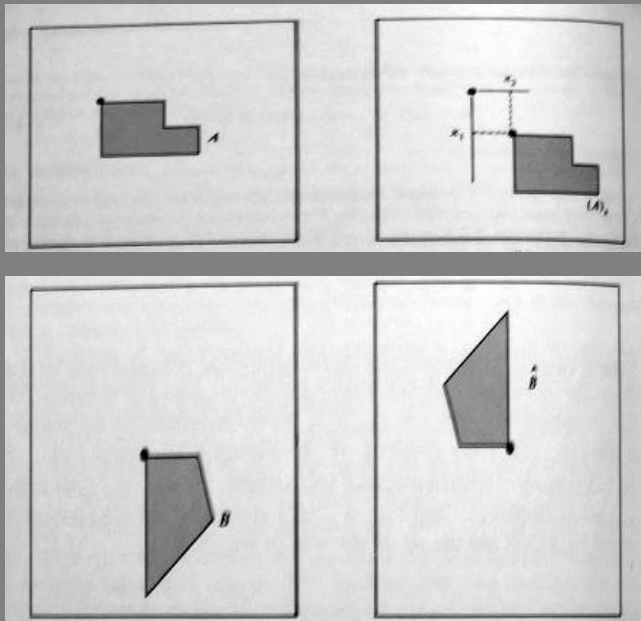


Image Preprocessing – Morphological Operations II

Dilation of A by B: $A \oplus B = \{x \mid [(\hat{B})_x \cap A] \neq f\}$

Erosion of A by B: $A \ominus B = \{x \mid (B)_x \subseteq A\}$

Note $(A \ominus B)^c = A^c \oplus \hat{B}$

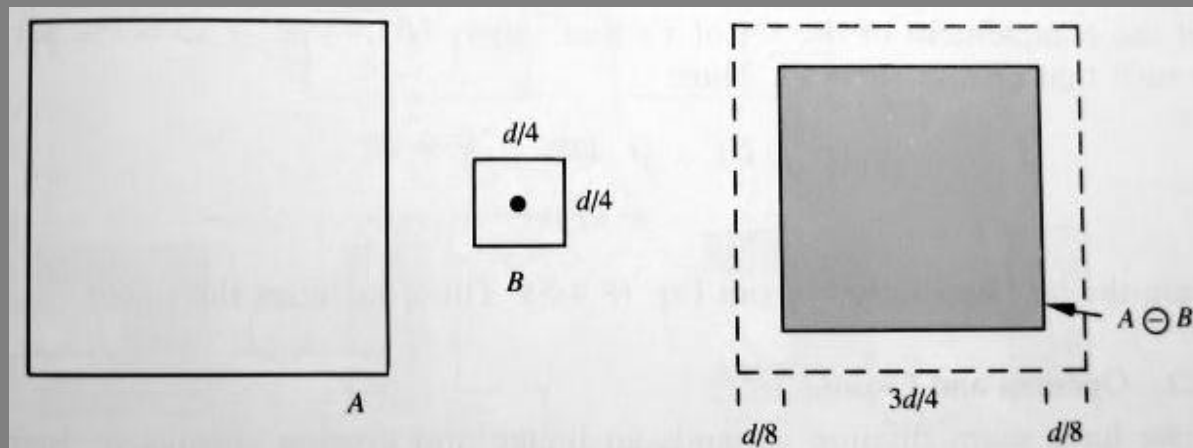
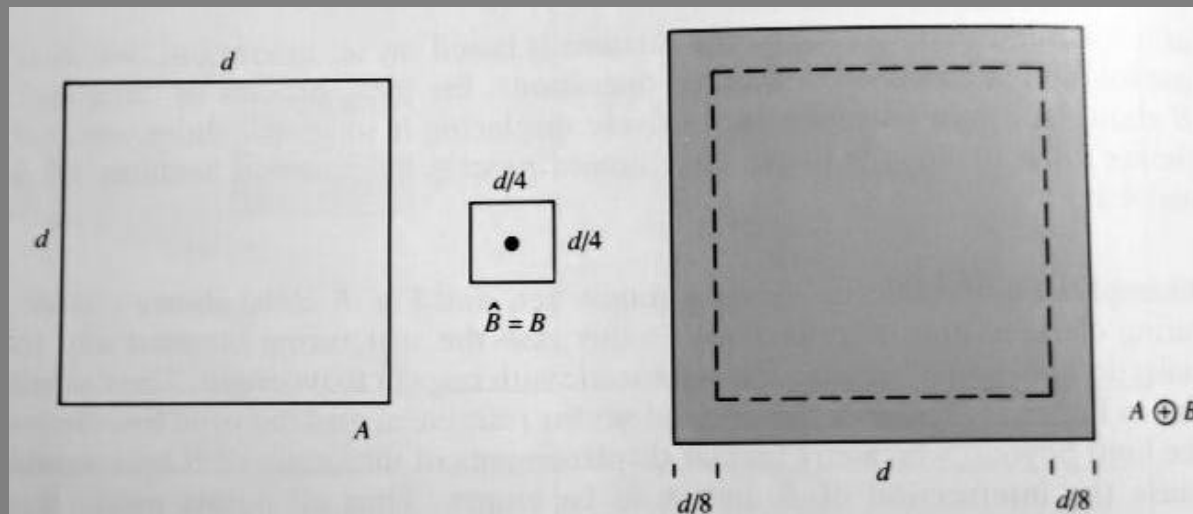


Image Preprocessing – Morphological Operations III

Opening of A by B: $A \circ B = (A \ominus B) \oplus B$

Closing of A by B: $A \bullet B = (A \oplus B) \ominus B$

Note $(A \circ B) \circ B = A \circ B$ and $(A \bullet B) \bullet B = A \bullet B$

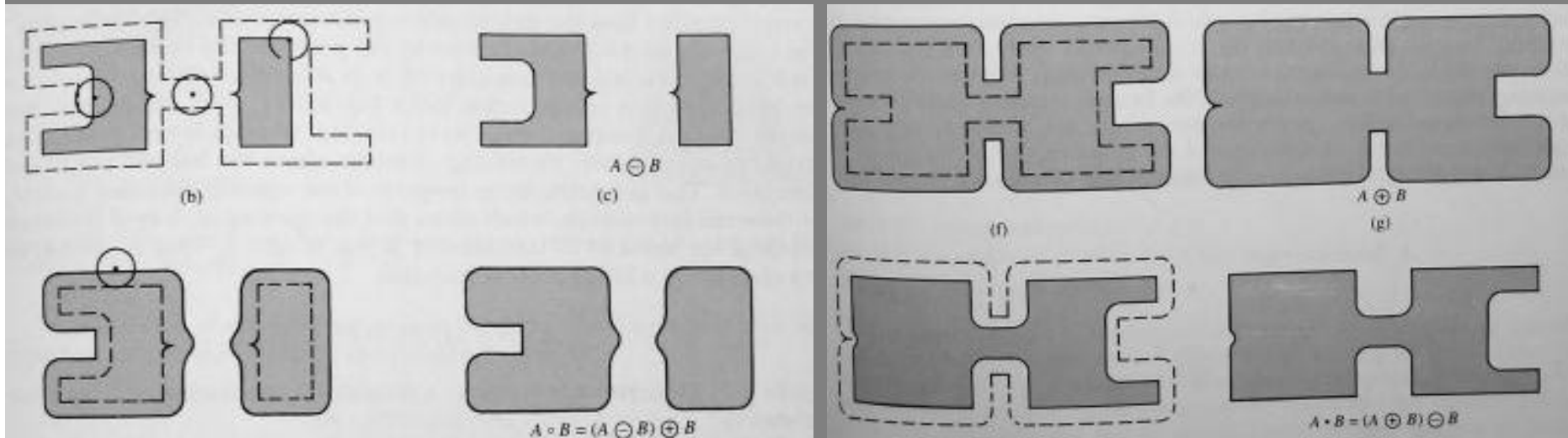
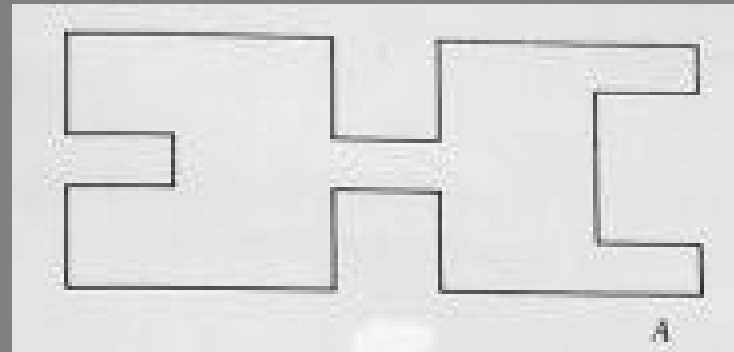
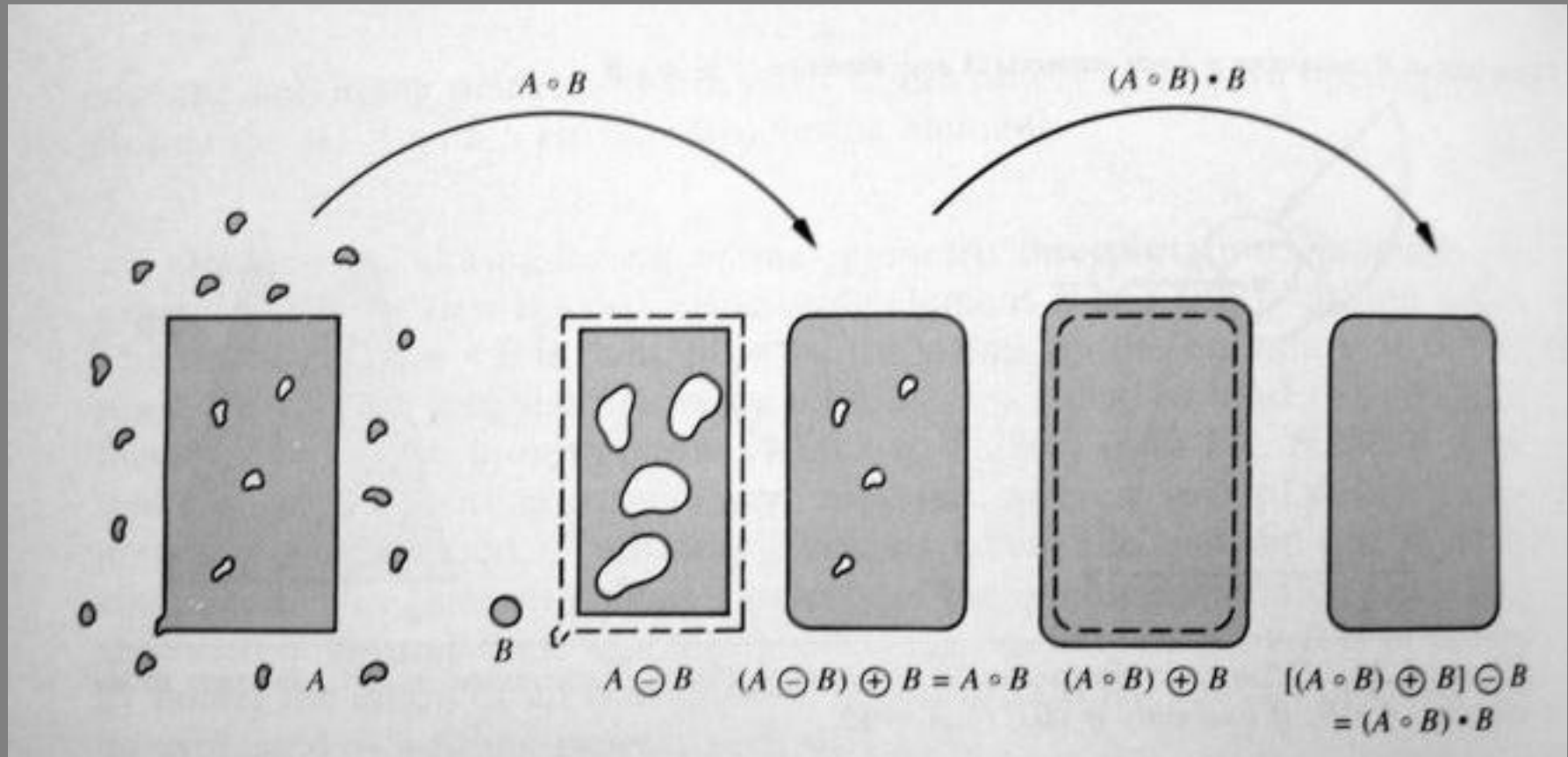


Image Preprocessing – An application of Opening/Closing

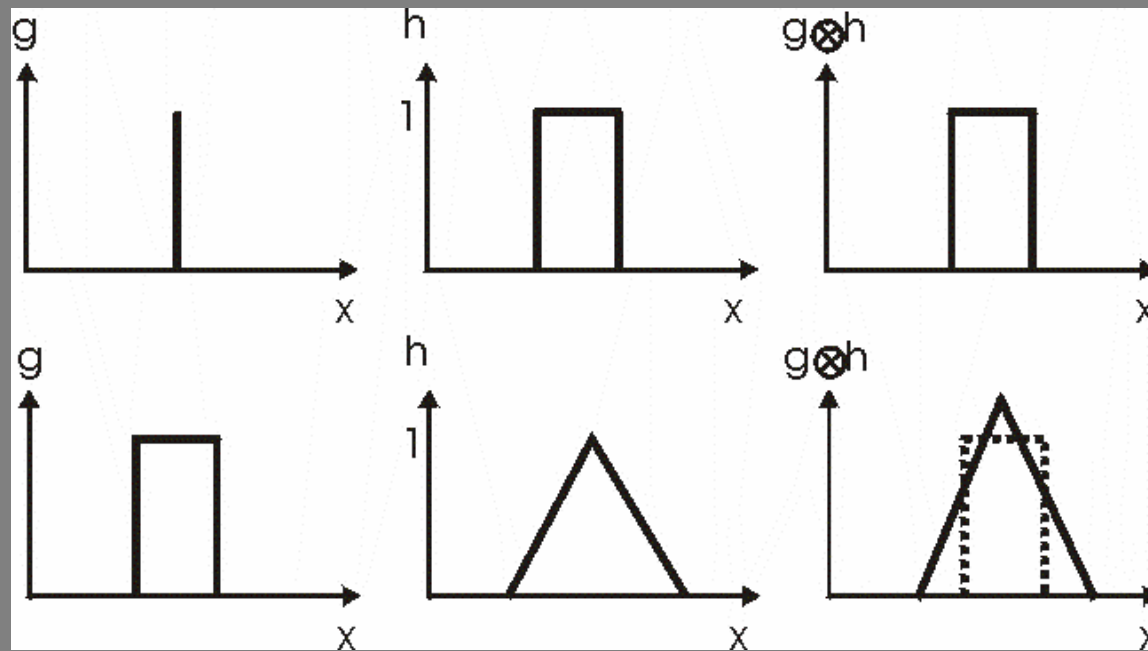


Convolution and Image Processing

Recall the definition of convolution:

$$g(t) \otimes h(t) = \int_{-\infty}^{\infty} g(\mathbf{t})h(t-\mathbf{t})d\mathbf{t}$$

Graphical explanation of convolution:



Convolution Theorem

$$\mathfrak{I}(g \otimes h)(f) = \tilde{g}(f)\tilde{h}(f)$$

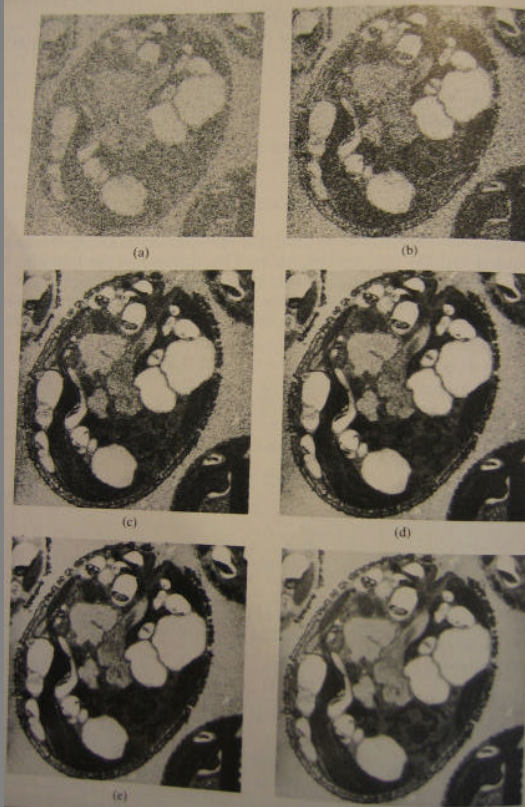
$$\begin{aligned}\int_{-\infty}^{\infty} g \otimes h(t) e^{-i2\mathbf{p}ft} dt &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(\mathbf{t}) h(t - \mathbf{t}) d\mathbf{t} e^{-i2\mathbf{p}ft} dt \\ &= \int_{-\infty}^{\infty} d\mathbf{t} g(\mathbf{t}) e^{-i2\mathbf{p}ft} \left(\int_{-\infty}^{\infty} dt h(t - \mathbf{t}) e^{-i2\mathbf{p}f(t - \mathbf{t})} \right) \\ &= \int_{-\infty}^{\infty} d\mathbf{t} g(\mathbf{t}) e^{-i2\mathbf{p}ft} \left(\int_{-\infty}^{\infty} dt' h(t') e^{-i2\mathbf{p}f(t')} \right) \\ &= \tilde{g}(f) \tilde{h}(f)\end{aligned}$$

where $t' = t - \mathbf{t}$ $dt' = dt$

Image Preprocessing – Noise Reduction, averaging, low pass filter, median filter

Averaging M images reduce noise by \sqrt{M}

$$S_M = \frac{1}{\sqrt{M}} S_1$$



Avg by 2,8,16,32,128

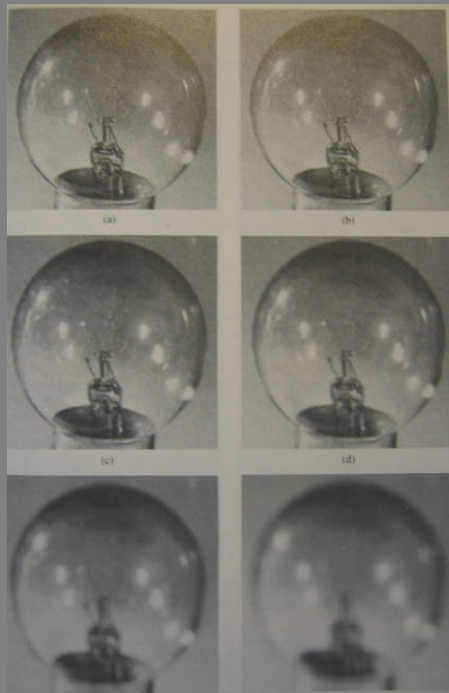
Low Pass Filter

3x3 kernel

$$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

5x5 kernel

$$\frac{1}{25} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$



Kernel 3,5,7,15,25

Median Filter

Replace center pixel value by the median value from a nxn pixel neighborhood

$$\begin{bmatrix} 7 & 12 & 10 \\ 6 & 9 & 12 \\ 8 & 10 & 200 \end{bmatrix}$$

Avg = 30; Median = 10



5x5 low pass vs median

Image Preprocessing – Fourier Filtering I

2D Fourier Transform

$$F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \exp[-2\pi i(ux + vy)] dx dy$$

Power Spectrum

$$P(u, v) = |F(u, v)|^2$$

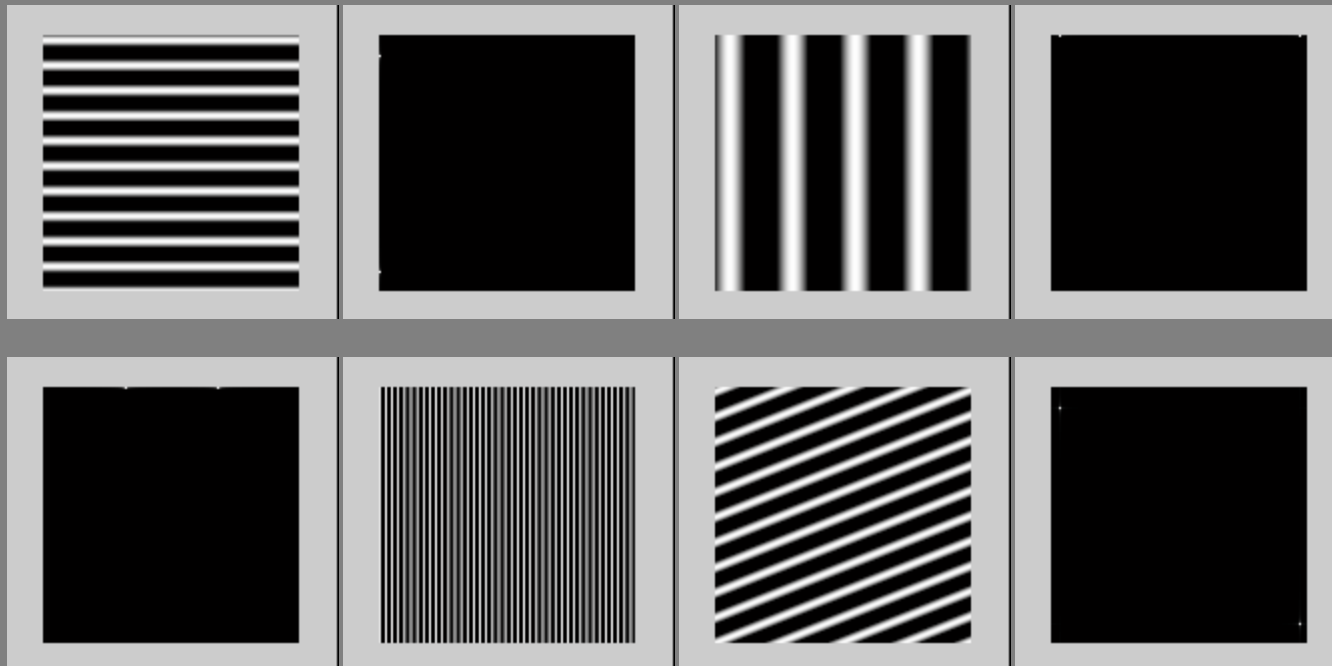
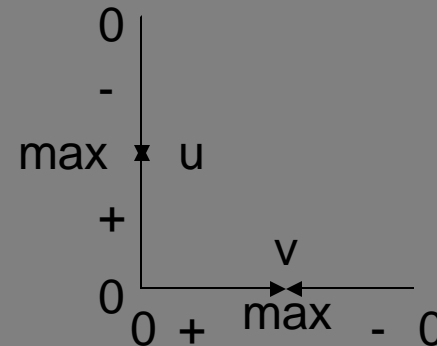
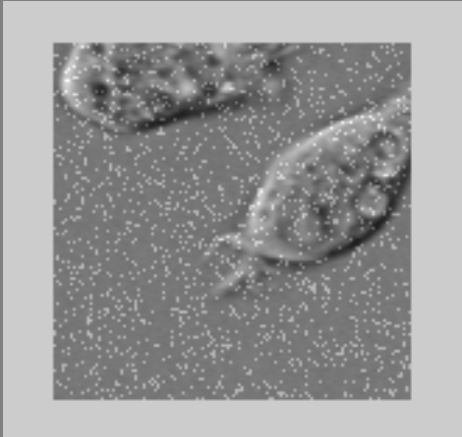


Image Preprocessing – Fourier Filtering II

Noise added



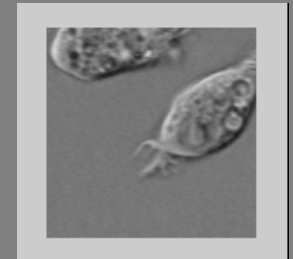
FFT low pass



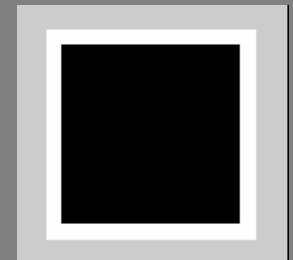
Convolution $\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$



original



FFT mask



VAR= 100

41

33

Image Preprocessing – Fourier Filtering III

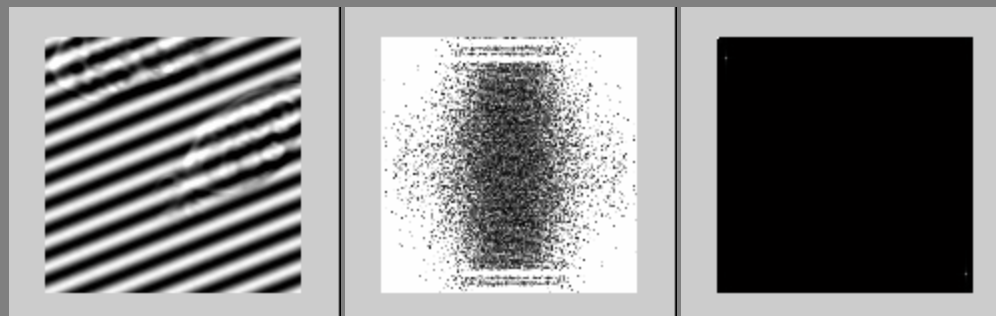
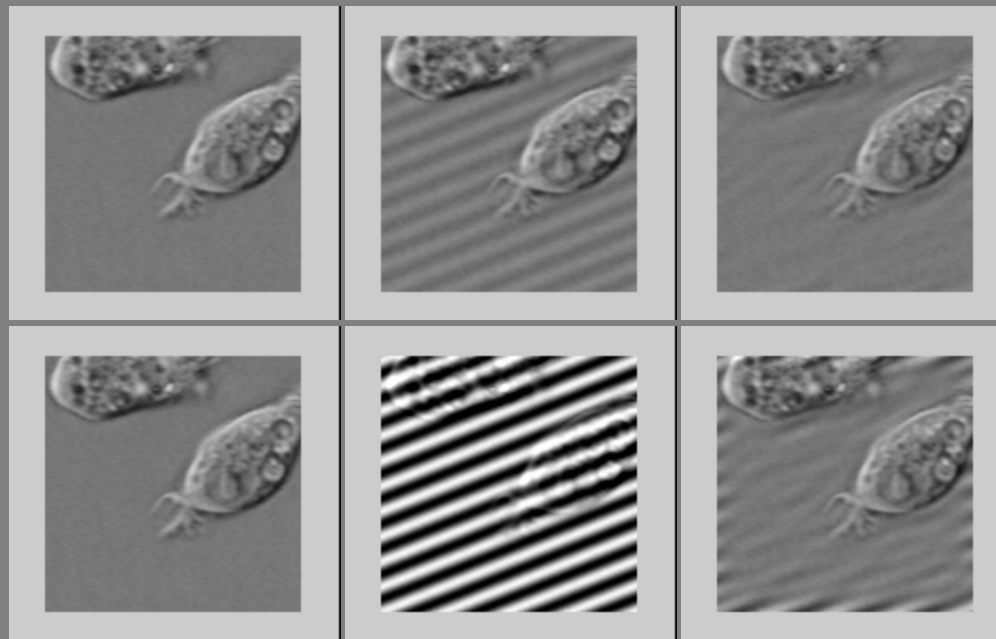
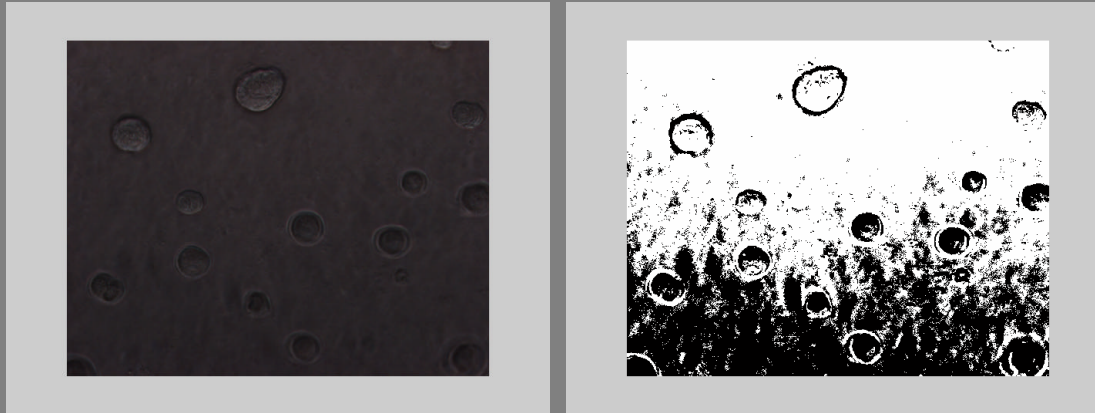


Image Segmentation – Global Threshold, Optimal Threshold

Simple global threshold:
$$N(i, j) = \begin{cases} 1 & \text{if } O(i, j) > T \\ 0 & \text{if } O(i, j) \leq T \end{cases}$$



Optimal thresholding

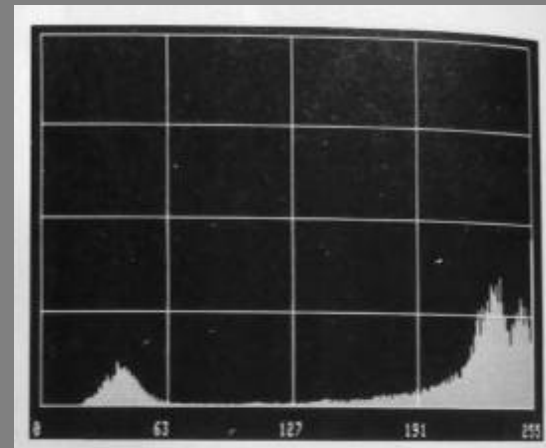
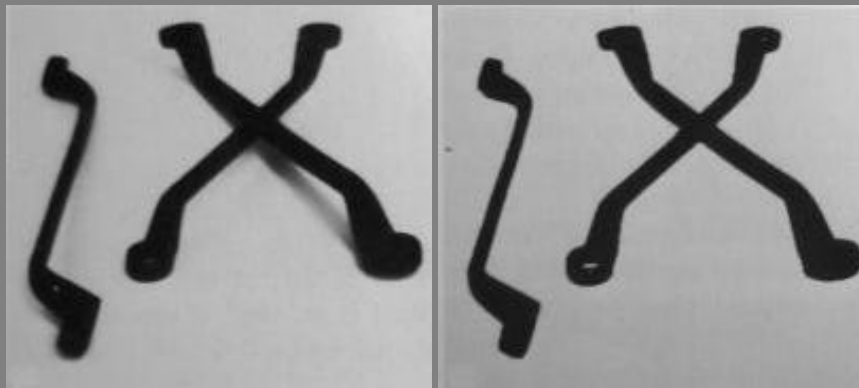


Image Segmentation – Edge Detection I

The edge is the boundary of two region with distinct intensity levels

The basic idea of edge detection is to compute the local derivative operator

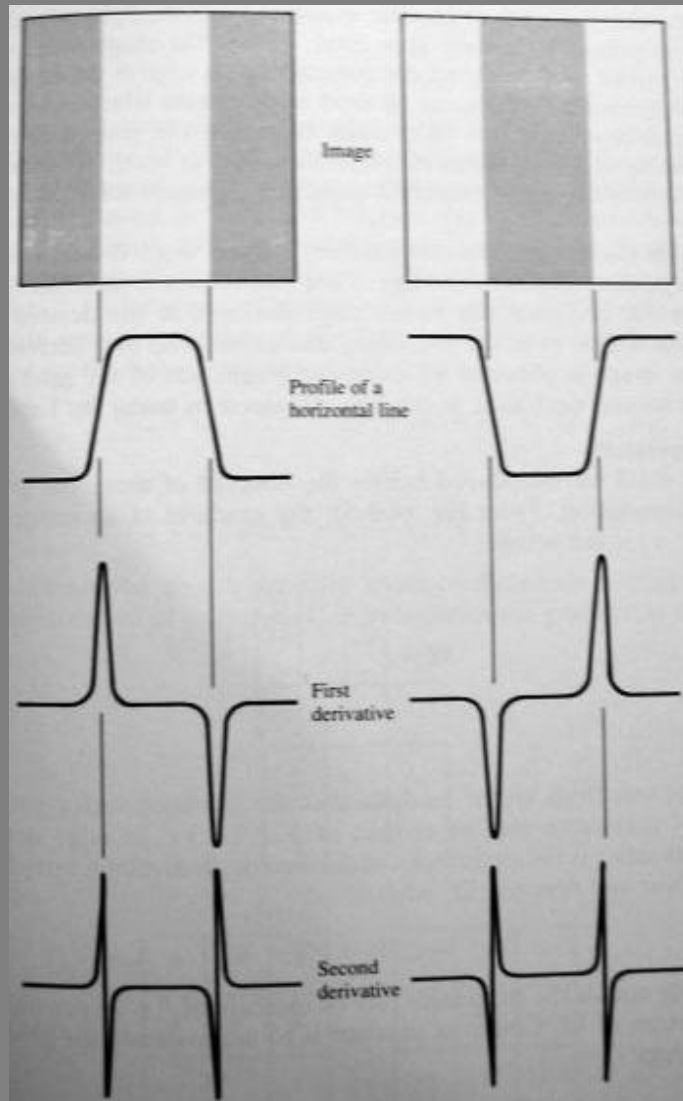


Image Segmentation – Edge Detection II

Gradient at each point (x,y) of an image

$$\nabla \vec{f} = \begin{bmatrix} G_x \\ G_y \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix}$$

Magnitude $\nabla f = \text{mag}(\nabla \vec{f}) = (G_x^2 + G_y^2)^{1/2} \approx |G_x| + |G_y|$

Direction $\mathbf{a} = \tan^{-1}\left(\frac{G_y}{G_x}\right)$

Sobel Implementation

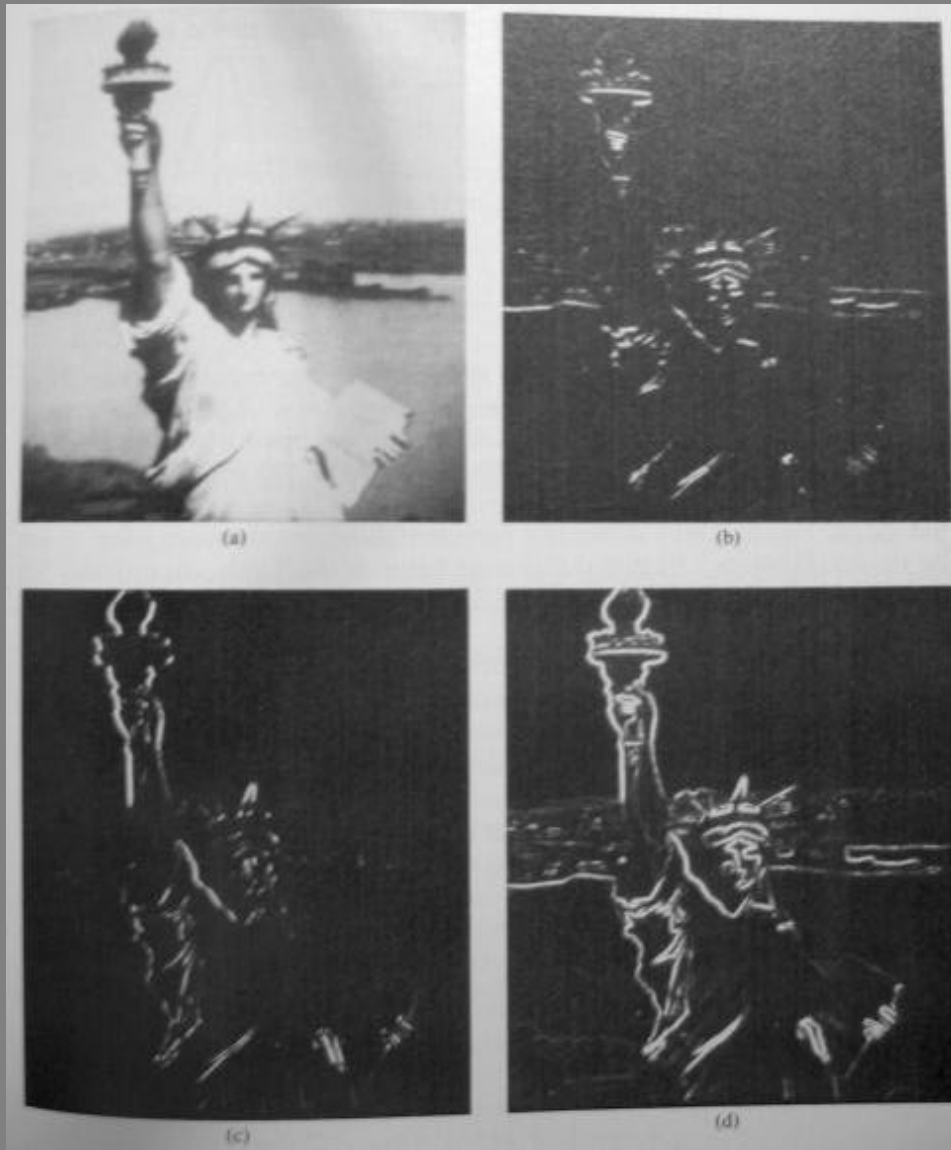
$$G_x = (z_3 + 2z_6 + z_9) - (z_1 + 2z_4 + z_7) \quad G_y = (z_7 + 2z_8 + z_9) - (z_1 + 2z_2 + z_3)$$

$$\begin{bmatrix} z_1 & z_2 & z_3 \\ z_4 & z_5 & z_6 \\ z_7 & z_8 & z_9 \end{bmatrix} \quad G_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} \quad G_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

Laplacian at each point (x,y) of an image

$$\nabla^2 f = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} = 4z_5 - (z_2 + z_4 + z_6 + z_8)$$

Image Segmentation – Edge Detection III



G_y

G_x

$$|G_x| + |G_y|$$

Image Segmentation – Boundary based Threshold and Region Filling

(1) Identify region based on first finding the boundaries:

$$s(x, y) = \begin{cases} 0 & \text{if } \nabla f < T \\ + & \text{if } \nabla f > T \text{ and } \nabla^2 f \geq 0 \\ - & \text{if } \nabla f > T \text{ and } \nabla^2 f < 0 \end{cases}$$

Creates a 3 level image based on gradient and Laplacian operators.
Boundaries at transition of (-,+) and (+,-)

(2) Edge Linking

After boundary pixels are identified. Due to noise, a linking procedure is often needed. Linking can be accomplished by closing operation or regional growing algorithms (discussed later)

